

# **Estimates of Wind Energy Input to the Ekman Layer in the Southern Ocean from Surface Drifter Data**

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**Abstract.** The energy input to the upper ocean Ekman layer is assessed for the Southern Ocean by examining the rotary cross-spectrum between wind stress and surface velocity from frequencies between 0 and 2 cpd. The wind stress is taken from ECMWF ERA40 reanalysis, and drifter measurements from 15 m depth are used to represent surface velocities, with an adjustment to account for the vertical structure of the upper ocean.

The combination of a stronger anticyclonic wind stress forcing associated with a greater anticyclonic response makes the contribution from the anticyclonic frequencies dominate the wind energy input.

The latitudinal and seasonal variations of the wind energy input to the Ekman layer are closely related to the variations of the wind stress, both for the mean and for the time-varying components. The contribution from the near-inertial band follows a different trend, increasing from 30° S to about 45° S and decreasing further south, possibly a consequence of the lack of variance in this band in the drifter and wind stress data.

Uncertainties in the analysis may be as large as 30%, but are nonetheless sufficient to show the relative importance of time-averaged, subinertial, and near-inertial wind forcing in transferring energy from the atmosphere to the ocean at the local scale. The high frequency forcing of the ocean is related to the sources of kinetic energy and near-inertial waves that are in turn related to the mixing of the ocean and the global overturning circulation.

## 1. Introduction

### 1.1. A spectral view of the wind energy input

The transfer of mechanical energy from the atmosphere to the ocean takes place through various physical mechanisms that can be distinguished by the time scales on which they occur. This study makes use of surface drifter data to examine the wind energy input to the Ekman layer on time scales ranging from low-frequency 40-day periods to near near-inertial frequencies.

First, at low frequencies, the wind stress  $\boldsymbol{\tau}$  works on the ocean general circulation (represented by the geostrophic velocity  $\mathbf{u}_g$  at the surface) as:

$$\boldsymbol{\tau} \cdot \mathbf{u}_g \equiv \mathbf{U}_e \cdot \nabla p_s / \rho. \quad (1)$$

Eq. (1) can be seen as either a direct generation of geostrophic kinetic energy or as an increase in potential energy by the work of the Ekman transport  $\mathbf{U}_e$  against the pressure forces at the surface  $\nabla p_s / \rho$  [Gill *et al.*, 1974; Fofonoff, 1981], with  $\rho$  the water density.

Weijer and Gille [2005] showed that the potential energy framework is more germane to the energetics of a numerical model of the Southern Ocean. Wunsch [1998] estimated the left-hand side of Eq. (1) by using altimeter-derived geostrophic velocities and wind stress field analyses for the 1992-96 time period. In his estimate, 95% of wind energy input came from the time mean components of  $\boldsymbol{\tau}$  and  $\mathbf{u}_g$ . Over the global ocean, the maximum energy input rates were greater than  $20 \times 10^{-3} \text{ W m}^{-2}$  and were found in the Southern Ocean (his Fig. 2). Moreover, 70% of the wind work integrated over the surface of the global ocean was found south of  $40^\circ \text{ S}$ . Huang *et al.* [2006] studied the decadal variability of this energy input from both altimetry data and numerical model outputs. They found that

the ACC region is where most of the variability is found for the 1979-2003 time period and furthermore that it is increasingly dominant in the global integral. More recent work has taken into account the influence of the current on the atmospheric stress [Duhaut and Straub, 2006; Hughes and Wilson, 2008]. Using extended observations, Hughes and Wilson [2008] updated Wunsch's estimates, reaching the same general conclusions, albeit with a decrease in the global integral because of the definite negative contribution to the wind work due to the effect of oceanic currents on atmospheric wind stress.

Second, at high frequencies, near-inertial motions are triggered in the mixed layer by rapid, small-scale wind stress fluctuations [e.g. D'Asaro, 1985b; Poulain, 1990]. Resonance can occur when the wind stress itself contains a significant rotary spectral component at the inertial frequency [e.g. Crawford and Large, 1996]. The energy flux at near-inertial frequencies has attracted interest, because it is potentially oriented downward towards the ocean interior by propagating internal waves [e.g. D'Asaro et al., 1995]. These waves eventually dissipate and can drive diapycnal mixing [Gregg, 1987]. The energy flux to mixed-layer near-inertial motions has been estimated from wind stress and ocean velocities measured at mooring locations [D'Asaro, 1985a] or more recently from global wind stress analyses [Alford, 2001; Watanabe and Hibiya, 2002; Alford, 2003]. In these global studies, the inertial velocities  $\mathbf{u}_i$  were inferred from Pollard and Millard's (1970) damped slab-layer model for which the inertial currents are uniform in the vertical. By using either numerical integration in time or spectral resolution of the equations, the wind energy input was computed as  $\boldsymbol{\tau} \cdot \mathbf{u}_i$ . Alford [2003] also considered depth-uniform Ekman velocities  $\mathbf{u}_e$  arising from a time-varying Ekman transport, because their energy is also available for dissipation, but found their impact to be modest. For the period 1989-95, the average

zonal-mean energy input rate at  $40^\circ$  S was about  $3 \times 10^{-3} \text{ W m}^{-2}$  (his Fig. 1). However, this value may be too large since *Plueddemann and Farrar* [2006] showed that the slab layer model systematically results in an overestimation of the work done on mixed-layer inertial currents, largely because a slab model lacks dissipation on short time scales.

Third, energy fluxes seem to take place at intermediate or “sub-inertial” frequencies through Ekman dynamics. Vertically-sheared Ekman currents are expected to exist for all forcing frequencies according to *Gonella* [1972] “extensions” of *Ekman’s* (1905) theory. He first derived several transfer functions that provide a mathematical representation of the wind-driven velocity as a function of depth and forcing frequency. *Elipot and Gille* [2008] studied the theoretical linear transfer functions arising from nine variants of the Ekman model, which differed in the parameterization of the vertical eddy viscosity and the boundary condition at the bottom on the wind-driven layer. Although these transfer functions vary in the details, in general, the predicted Ekman velocities  $\mathbf{u}_e$  spiral with depth from the surface and rotate along with the forcing wind stress vector. The transfer functions all predict a stronger response when the frequency of the forcing is anticyclonic (counterclockwise or positive in the Southern Hemisphere) than when it is cyclonic. They also exhibit either an unbounded or a maximum resonant response at the inertial frequency. While the Ekman currents themselves should play a role in the wind energy input into geostrophic currents via Eq. (1), maintaining the Ekman spiral requires dissipation within the Ekman layer, or a wind energy input rate “to the Ekman layer”. *Elipot and Gille’s* (2008) results showed that the best models to explain drifter observations in the Southern Ocean have a bounded response at the inertial frequency

at which there is a zero cross-wind component of the wind-driven velocity, and the wind energy input is therefore maximal.

*Wang and Huang* [2004] estimated the wind energy input to the Ekman layer by using the transfer function to compute the ocean surface Ekman velocities  $\mathbf{u}_e$  from wind stress analyses, assuming that the Ekman model with a constant vertical viscosity and infinite depth ocean applied. They set the Ekman depth to be proportional to the mean friction velocity computed from the wind stress and inversely proportional to the Coriolis parameter. The total wind energy input into the Ekman layer was then obtained by summing the dot product  $\boldsymbol{\tau} \cdot \mathbf{u}_e$  over all frequency components. They found that the global integral for this energy input is also dominated by the contribution from the Southern Hemisphere. Examination of Fig. 3 in *Wang and Huang* [2004] reveals that this is mainly due to the strong winds over the Southern Ocean and notably large values over the ACC of about  $20 \times 10^{-3} \text{ W m}^{-2}$ . They also computed the changes of the global energy input to the Ekman layer over the 1948-2002 time period, which *Huang et al.* [2006] subsequently attributed to wind stress variability in the equatorial region and in the Southern Ocean.

## 1.2. Does the wind energy input into the Ekman layer matter?

We will show in this paper that the observed co-spectrum between wind stress and oceanic velocities as measured by drifters at 15 m is representative of a momentum flux from the atmosphere to the ocean. This flux takes place at a wide range of frequencies, from low to “sub-inertial” and inertial frequencies. The wind work on the geostrophic circulation is actually a source of mechanical energy for the interior ocean [*Stern, 1975*], but previous studies have not resolved how the wind works on ageostrophic motions, nor have they explained how the generation of ageostrophic motions leads to a transfer of

114 mechanical energy to the ocean interior, well below the surface [*Ferrari and Wunsch*,  
115 2008; *Plueddemann and Farrar*, 2006]. Thus one may ask if the third pathway for energy  
116 input described above, due to subinertial oscillations, is a relevant source for the energy  
117 needed to sustain global meridional overturning circulation.

118 *Von Storch et al.* [2007] derived a mechanical energy budget for output from a high-  
119 resolution general circulation model that represented vertical mixing with the nonlocal  
120 K profile Parameterization (KPP) of *Large et al.* [1994]. Their analysis showed that the  
121 power provided by the wind at the sea surface was not separated between the work done  
122 on the geostrophic and ageostrophic circulation, because both components contributed to  
123 the total energy transferred down the water column. The dominant terms in the energy  
124 equation appear to be the pressure work and the shear-induced stress but the relative  
125 importance of these two terms was seen to vary as a function of depth. Overall, their  
126 results suggested that about 30% of the total wind energy input, through geostrophic and  
127 ageostrophic pathways, is passed through the interior ocean, while the remaining 70%  
128 is dissipated locally or converted to potential energy in the upper ocean. These results  
129 suggested that it is therefore necessary to estimate all pathways of wind energy input at  
130 the surface of the ocean.

131 *Von Storch et al.* [2007] did not distinguish the inertial band contribution from the  
132 the contribution due to Ekman dynamics at other frequencies. This is understandable,  
133 because the concept of a slab layer at near-inertial frequencies is incompatible with the  
134 full primitive equations employed in an OGCM. Moreover, their model forcing had 1-  
135 day resolution and may have suppressed higher-frequency variability. Indeed, theoretical  
136 analysis of internal wave generation by surface forcing [*Gill*, 1984], a recent numerical

large eddy simulation [*Polton et al.*, 2008], and observations [*D’Asaro et al.*, 1995; *Alford and Whitmont*, 2007] all have shown that near-inertial processes are important in the ocean.

### 1.3. Estimates from in-situ data in this study

In this study, we show how to isolate and estimate the wind energy input in the Ekman layer from in-situ data. This approach merits attention, if only because the concept of wind energy input to the Ekman layer has led to several estimates previously mentioned [*Wang and Huang*, 2004; *Huang et al.*, 2006] but none from in-situ observations of oceanic velocities on large scales. At least locally in the Southern Ocean, it complements other data-based estimates of the wind work on geostrophic motions. We here focus on the Southern Ocean (rather than the global ocean), because as previous studies have noted, it is the primary region for wind forcing to drive the ocean, and this wind energy is therefore thought to contribute to wind-induced mixing needed to sustain the thermohaline structure of the world ocean [*Wunsch and Ferrari*, 2004].

The steps leading to the development of the equations in this paper, their interpretation and the subsequent estimates were motivated by observations of coherent sub-inertial motions in drifter data [*Weller*, 1981; *McNally et al.*, 1989; *Niiler and Paduan*, 1995; *Rio and Hernandez*, 2003; *Elipot*, 2006]. This correlation as a function of frequency has been explained quite clearly by the development of Ekman theory extensions [*Gonella*, 1972; *Rudnick and Weller*, 1993; *Elipot and Gille*, 2008]. In this paper we derive a spectral energy equation and show that the wind energy input to the Ekman layer (i.e. the energy input in the oceanic boundary layer as represented by “Ekman” models) is equal to the real part of the cross-spectrum between the ocean surface velocity and the atmospheric



wind stress (section 2). First the cross-spectrum at the surface from surface drifter data and reanalysis wind stress (section 3) is estimated; however, because drifter velocities are not exactly representative of currents at the surface but at 15 m, some knowledge of the vertical structure of the transfer function is needed to obtain the cross-spectrum at the surface. The results from *Elipot and Gille* [2008] are used to infer the cross-spectrum at the surface and therefore to estimate the wind energy input to the Ekman layer zonally-averaged in the Southern Ocean (section 4). The results are discussed in terms of spectral decomposition as well as latitudinal and seasonal variations. Section 5 provides a summary.

## 2. Spectral energy equation

The cross-spectrum between wind stress and ocean surface velocity serves as the primary diagnostic tool for this paper. It can be obtained theoretically from the horizontal momentum balance. The derivation of a spectral energy equation shows that the real part of the cross-spectrum, the co-spectrum, gives an estimate of the wind energy input rate to the Ekman layer.

### 2.1. Spectral energy equation: the balance of spectra

The linearized horizontal momentum balance for the ocean in the absence of horizontal pressure gradients is:

$$\frac{\partial \mathbf{u}(t, z)}{\partial t} + i f \mathbf{u}(t, z) = \frac{1}{\rho} \frac{\partial \boldsymbol{\tau}(t, z)}{\partial z}, \quad (2)$$

where here  $\mathbf{u}$  is the horizontal velocity,  $z$  the depth negative downward,  $\boldsymbol{\tau}$  is the turbulent Reynolds stress. The finite Fourier transform  $\int_0^T (\cdot) \exp(-i2\pi\nu_k t) dt$  of Eq. (2) is:

$$i(2\pi\nu_k + f) \mathbf{U}_k(\nu, z) = \frac{1}{\rho} \frac{\partial \mathbf{T}_k(\nu_k, z)}{\partial z}, \quad (3)$$

where  $\mathbf{T}_k$  is the Fourier transform of  $\boldsymbol{\tau}$ . Multiplying Eq. (3) by  $\mathbf{U}_k^*$  and integrating from  $z < 0$  to the surface  $z = 0$  gives:

$$i\rho(2\pi\nu_k + f) \int_z^0 |\mathbf{U}_k|^2 dz = \int_z^0 \frac{\partial \mathbf{T}_k}{\partial z} \mathbf{U}_k^* dz. \quad (4)$$

Integrating by parts the right hand side yields:

$$i\rho(2\pi\nu_k + f) \int_z^0 |\mathbf{U}_k|^2 dz = \mathbf{T}_k \mathbf{U}_k^* \Big|_{z=0} - \mathbf{T}_k \mathbf{U}_k^* \Big|_z - \int_z^0 \mathbf{T}_k \left( \frac{\partial \mathbf{U}_k}{\partial z} \right)^* dz. \quad (5)$$

The first term on the right-hand side of Eq. (5) corresponds to the wind stress boundary condition at the surface. The second term vanishes when we let  $z$  go to  $-\infty$  to represent the base of the wind-driven layer where velocities and/or turbulent stresses vanish, and hence their Fourier transforms go to zero. Next, we apply to this equation the expected value operator and divide by the length of the observation  $\mathcal{T}$  to form a spectral energy equation:

$$i(2\pi\nu_k + f)E + D = S_{\boldsymbol{\tau}\mathbf{u}}, \quad (6)$$

where

$$E = \rho \int_{-\infty}^0 \frac{\langle |\mathbf{U}_k|^2 \rangle}{\mathcal{T}} dz, \quad (7)$$

$$D = \int_{-\infty}^0 \frac{\langle \mathbf{T}_k \left( \frac{\partial \mathbf{U}_k}{\partial z} \right)^* \rangle}{\mathcal{T}} dz, \text{ and} \quad (8)$$

$$S_{\boldsymbol{\tau}\mathbf{u}} = \frac{\langle \mathbf{T}_k \mathbf{U}_k^* \Big|_{z=0} \rangle}{\mathcal{T}} \quad (9)$$

are estimates of the vertically integrated kinetic power spectral density of the Ekman layer, the vertically integrated dissipated power spectral density in the Ekman layer, and the power cross-spectrum between the wind stress and the ocean surface velocity. This equation does not give the rate of change of energy but rather the average balance among various power spectral quantities.

Next, we assume that the turbulent stress  $\boldsymbol{\tau}$  is proportional to the vertical shear (following a classical diffusion model),

$$\frac{\boldsymbol{\tau}(t, z)}{\rho} = K(z) \frac{\partial \mathbf{u}(t, z)}{\partial z}. \quad (10)$$

The stress is aligned with the velocity vertical shear, implying that the vertical eddy viscosity  $K$  is real and independent of time. No additional assumptions are applied to  $K$ .

The finite Fourier transform of Eq. (10) is:

$$\frac{\mathbf{T}_k}{\rho} = K \frac{\partial \mathbf{U}_k}{\partial z}. \quad (11)$$

Substituting (11) into (8) gives

$$D = \int_{-\infty}^0 \rho K \frac{|\frac{\partial \mathbf{U}_k}{\partial z}|^2}{T} dz \quad (12)$$

which is a real and positive quantity. The power cross-spectrum can be decomposed classically into real and imaginary parts:

$$S_{\boldsymbol{\tau}\mathbf{u}} = C_{\boldsymbol{\tau}\mathbf{u}} - i Q_{\boldsymbol{\tau}\mathbf{u}}, \quad (13)$$

where  $C_{\boldsymbol{\tau}\mathbf{u}}$  is the coincident spectrum (co-spectrum) and  $Q_{\boldsymbol{\tau}\mathbf{u}}$  is the quadrature spectrum (quad-spectrum). Thus, Eq. (6) splits into two equations:

$$D = C_{\boldsymbol{\tau}\mathbf{u}}, \quad (14)$$

$$-(2\pi\nu + f)E = Q_{\boldsymbol{\tau}\mathbf{u}}. \quad (15)$$

Eq. (14) describes the throughput of energy in each frequency band: the dissipated energy in the Ekman layer (i.e. the energy sink) equals the energy source given by the co-spectrum of the Ekman velocity at the surface and the wind stress. Eq. (15) states that the kinetic energy of the Ekman layer can be obtained from the quad-spectrum.

These equations resemble the spectral equations for atmospheric motions derived in the time domain by *Chiu* [1970], who found that the spectral density of the kinetic energy is shaped by quadrature spectral quantities and that the Coriolis parameter plays a role, as is apparent in Eq. (15).

## 2.2. Total energy input rate to the Ekman layer

The Wiener-Khinchine theorem states that the frequency integral of the cross-spectrum is equal to the complex cross-correlation function at zero time lag  $\mathbf{R}_{\tau\mathbf{u}}(0)$  [e.g. *Bendat and Piersol*, 1986]:

$$\int_{-\infty}^{+\infty} S_{\tau\mathbf{u}}(\nu) d\nu = \mathbf{R}_{\tau\mathbf{u}}(0) \quad (16)$$

$$= \langle (\boldsymbol{\tau} - \langle \boldsymbol{\tau} \rangle)(\mathbf{u} - \langle \mathbf{u} \rangle)^* \rangle + \langle \boldsymbol{\tau} \rangle \langle \mathbf{u}^* \rangle. \quad (17)$$

For the real part this gives:

$$\int_{-\infty}^{+\infty} C_{\tau\mathbf{u}}(\nu) d\nu = \mathcal{R} [\langle (\boldsymbol{\tau} - \langle \boldsymbol{\tau} \rangle)(\mathbf{u} - \langle \mathbf{u} \rangle)^* \rangle] + \mathcal{R} [\langle \boldsymbol{\tau} \rangle \langle \mathbf{u}^* \rangle], \quad (18)$$

which states mathematically that the integrated co-spectrum equals the real part of the complex covariance of  $\boldsymbol{\tau}$  and  $\mathbf{u}$  plus the product of their time means. This does not imply that the co-spectrum is the frequency distribution of the total work done on the ocean by the wind stress. The spectral analysis of the latter quantity would include the work on geostrophic currents and other processes including nonlinear forcings and interactions between frequencies.

## 3. On obtaining the surface cross-spectrum from drifter data

### 3.1. Data

Ideally, the cross-spectrum discussed in the previous section would be computed using time series of wind-driven velocity at the surface and contemporaneous wind stress time

series, at fixed locations, on a regular grid over the world ocean. However, observations from moorings are too sparse to provide representative estimates on global scales, and their shallowest velocity measurement depths are usually too far from the surface. Instead, here we take advantage of velocity data derived from surface drifter trajectories. Drifter data are not ideal to obtain the cross-spectrum at the surface and our estimates suffer from non-negligible uncertainties as well as some biases (see below). Nonetheless, we show here that a first estimate from data of the wind energy input to the Ekman layer in this region can be obtained and used as a benchmark for other indirect estimates. Such an estimate may eventually be refined as our understanding of Southern Ocean boundary layer dynamics improves.

To estimate the cross-spectrum, we use 40-day time series of wind stresses and ageostrophic velocities. The latter were computed by subtracting surface geostrophic velocities (see below) from drifter velocities. Drifter motions represent the currents averaged over the 6.1 m length of a drogue, centered at 15 m depth. Therefore we interpret the drifter velocities to be representative of oceanic velocities at 15-m depth, disregarding the vertical shear over the length of the drogue [Niiler *et al.*, 1995].

Geostrophic velocities were derived from gradients of 7-day altimetric sea surface height anomalies [AVISO], to which we added a time-mean geostrophic velocity computed from GRACE satellite-derived dynamic topography [Tapley *et al.*, 2005]. Here we find that the correlation between AVISO data and ECMWF data is not significant (not shown) when these two datasets are interpolated along drifter trajectories. [In contrast, Hughes and Wilson [2008] showed statistically significant correlation between wind stress and altimeter-derived geostrophic velocities in the regions of the ocean comprising continental

slopes, the Equatorial region and extensions of western boundary currents.] Although the AVISO/ECMWF correlation is not significant, the geostrophic component still must be removed from the drifter velocity before computing the cross-spectrum. This is because the integrated co-spectrum has a component arising from the time-mean components of the drifter velocity and the wind stress velocity (second term on the left-hand side of Eq. 17) that we would like to estimate as accurately as possible. Geostrophic components left in the mean or anomaly fields could lead to an erroneous estimate of the mean ageostrophic velocity.

For wind data, we use European Center for Medium-Range Weather Forecasts (ECMWF) ERA-40 Project re-analysis wind stresses [*Simmons and Gibson, 2000*] obtained from the Data Support Section of the Scientific Computing Division at the National Center for Atmospheric Research. ECMWF wind stresses are released on a Gaussian grid with resolution of  $1.125^\circ$  longitude by roughly  $1.125^\circ$  latitude at 6 hour sampling. These grids were linearly interpolated on the drifter positions to obtain contemporaneous time series of wind stress. Using drifter velocity time series without taking into account their Lagrangian nature is equivalent to assuming horizontal homogeneity when writing the momentum equation (2). This assumption is justified by the fact that the spatial length scales of the local wind stress events driving Ekman or inertial currents are larger than the length scales of the drifter motions. As in *Elipot and Gille* [2008], spectral estimates by the periodogram method and formal errors were computed following *Bendat and Piersol* [1986]. See *Elipot and Gille* [2008] for more details.

### 3.2. The shear bias

Drifter velocities are interpreted as 15 m velocities so that the cross-spectrum with the wind stress is not the quantity that we seek to estimate (Eq. 9) and a correction for this shear bias needs to be implemented. At subinertial frequencies, a model with vertical shear for wind-driven velocity is preferable to a slab model [e.g. *Weller and Plueddemann, 1996; Elipot and Gille, 2008*]. In the spectral domain, the transfer function theoretically links the auto-spectrum of the wind stress to the cross-spectrum of the wind stress and the velocity at depth  $z$  [*Elipot and Gille, 2008*]:

$$S_{\tau u}(\nu, z) = \mathbf{H}(\nu, z) S_{\tau \tau}(\nu). \quad (19)$$

The real part of (19) is:

$$C_{\tau u}(\nu, z) = \mathcal{R} [\mathbf{H}(\nu, z)] S_{\tau \tau}(\nu) \quad (20)$$

Applying this expression for  $z = 0$  and  $z = -15$ , the depth of a drifter's drogue, the co-spectrum of the wind stress and the ocean velocity is:

$$C_{\tau u}(\nu, 0) = \frac{\mathcal{R} [\mathbf{H}(\nu, 0)]}{\mathcal{R} [\mathbf{H}(\nu, -15)]} C_{\tau u}(\nu, -15) \quad (21)$$

$$= R \times C_{\tau u}(\nu, -15). \quad (22)$$

$R$  is a shear correction coefficient that can be derived from the transfer function  $\mathbf{H}$ . *Elipot and Gille* [2008] estimated the transfer function from the same drifter and wind stress data used in this study. They fitted these observed transfer functions to analytic transfer function that used different expressions for the oceanic boundary layer and for vertical viscosity and then sought best estimates of viscosity and boundary layer parameters in order to minimize misfit between observations and theory. They found that the overall best model across the Southern Ocean is a 1-layer model with a constant viscosity  $O(0.01-0.1) \text{ m}^2 \text{ s}^{-1}$  and a boundary layer depth  $O(30-50) \text{ m}$ , with latitudinal variations. Here

we use the parameters from this model to compute  $R$  to obtain the co-spectrum at the surface by Eq. 21.

Theoretically,  $1/R$  is less than one and more convenient to plot. Fig. 1 shows  $1/R$  as a function of latitude and frequency from the results of *Elipot and Gille* [2008].

### 3.3. The slip bias

Surface drifters are slightly imperfect water-followers, and their measurements include an erroneous slip velocity thought to be caused by two phenomena: First, the direct action of the wind on the surface flotation buoy, and second the vertical shear of the horizontal velocity across the vertical extend of the drogue. *Niiler et al.* [1995] estimated the slip  $\mathbf{u}_s$  using vector measuring current meters mounted on the top and bottom of the drogues for a select group of drifters released in the tropical and northeastern Pacific. They modeled the slip as:

$$\mathbf{u}_s = \frac{a}{A} \mathbf{w}_{10} + \frac{b}{A} \Delta \mathbf{u}, \quad (23)$$

where  $\mathbf{w}_{10}$  is the 10-m wind velocity,  $A$  is the drag area ratio of the drogue to the other constituents of a drifter (40 for a SVP-type drifter) and  $\Delta \mathbf{u}$  is the velocity difference measured between the top and bottom of the drogue. Since  $\Delta \mathbf{u}$  cannot be determined from drifter data alone, here we do not attempt to evaluate the second term in (23). We correct drifter velocities for wind slip using  $a = 4.63 \times 10^{-2}$  [*Niiler et al.*, 1995] and determining  $\mathbf{w}_{10}$  from ECMWF 10-m winds interpolated in space and time to drifter locations. As noted by [*Niiler et al.*, 2003], the wind slip correction may be underestimated in the Southern Ocean, where mean wind speeds often exceed  $10 \text{ m s}^{-1}$ , the upper limit for which estimates of  $a$  have been validated.



The basic impact of this slip correction is to modify the estimates of the cross-spectrum, but it potentially acts at two steps of the wind energy input computation. First, *Elipot and Gille* [2008] found that applying the wind slip correction tends to increase the magnitude of the vertical viscosity and of the boundary layer depth. This in turn reduces the shear from the surface to 15 m depth. Second, the wind slip correction also reduces the magnitude of the co-spectrum and increases the quad-spectrum. Correspondingly, this reduces the inferred co-spectrum of the wind stress and the surface ocean velocity after applying the shear correction. Finally, the estimates for the energy input rate  $D$  are accordingly reduced. North of  $54^\circ$  S, the reduction averages 32%, but it is only 13% south of this latitude, equivalent to a 3 and  $17 \times 10^{-3} \text{ W m}^{-2}$  reduction across the Southern Ocean, with an average decrease of  $8.6 \times 10^{-3} \text{ W m}^{-2}$ . These uncertainties could seem deceptively large but are in fact reasonable for a first estimate of this type of wind energy input from in-situ data. It also clearly calls for further research into the dynamics of the boundary layer as well as the behavior of SVP drifters in high wind environments.

#### 4. Results and Discussion

To summarize the method, we compute the co-spectrum at 15 m, estimate the surface co-spectrum following Eq. 21, and then sum the surface co-spectrum over the frequency range resolved in this study (from -2 cpd to 2 cpd) to obtain the wind energy input following the left-hand side of Eq. 18. At this stage the reader may wonder why we do not simply compute the cross-covariance between  $\boldsymbol{\tau}$  and  $\mathbf{u}$ , that is the right-hand side of Eq. 18? The first reason is that the shear correction depends on the frequency of the motions (as shown clearly by Fig. 1) and cannot be applied to only the cross-covariance, as pointed out by *Crawford and Large* [1996]. Second, by retaining the frequency information

we gain a useful diagnostic of the frequencies that matter for the process under scrutiny,  
as will be shown in section 4.3.

In order to study latitudinal and seasonal variability, the total drifter dataset is divided  
in  $2^\circ$  latitudinal bands, without regard for dynamical regime. Co-spectra were estimated  
in each of these bands and integrated over specified frequency ranges. Figure 3 shows  
results for the year-round dataset, and Fig. 4 shows results for austral winter (blue) and  
summer (red).

#### 4.1. Year-round data

The total energy input rate is plotted in black in Fig. 3a. It differs by more than a factor  
of 5 across the Southern Ocean, ranging from a minimum of  $(9.8 \pm 0.3) \times 10^{-3} \text{ W m}^{-2}$   
at  $31^\circ \text{ S}$  to a maximum of  $(53 \pm 3) \times 10^{-3} \text{ W m}^{-2}$  at  $57^\circ \text{ S}$ . We split the total energy  
into contributions from the mean, or zero frequency, (light gray line) and contributions  
from the time-varying components (dark gray line). Both increase from north to south,  
the mean by more than a factor of 10 and the time-varying contribution by a factor of  
4. Fig. 3b shows that their relative contributions are comparatively consistent at all  
latitudes, with the mean having its most significant impact at  $51^\circ \text{ S}$ , where it accounts  
for 28% of wind energy input. Wind stress, shown in Fig. 3c is largest at  $53^\circ \text{ S}$  for the  
mean (light gray), the variance (dark gray), and the total (black).

Fig. 3d shows the energy input rate to the Ekman layer split into cyclonic and anticy-  
clonic frequencies. It is predominantly anticyclonic at almost all latitudes in the Southern  
Ocean, except southward of  $57^\circ \text{ S}$  where about an equal amount of energy is input through  
cyclonic and anticyclonic frequencies. The relative contributions of the anticyclonic and  
cyclonic frequencies are plotted in Fig. 3e: the anticyclonic contribution decreases from

72% at 31° S to 50% at 59° S. (For this plot, the contribution of the mean is partitioned equally between the two frequency domains.) This anticyclonic predominance is a consequence of two factors. First, as shown in Fig. 3f, the mean-squared wind stress forcing is larger for anticyclonic frequencies (dashed lines) than for cyclonic frequencies (solid lines) across the Southern Ocean. [The mean-squared wind stress  $\langle \tau^2 \rangle$  is about 70% anticyclonic frequencies at 31° S, decreasing to about 53% anticyclonic at 59° S (not shown).] Second, if the oceanic response were flat, then the ocean currents would exactly reflect the partitioning of the wind. Instead, here the ocean response is more strongly anticyclonic, due to the preferential anticyclonic response of the ocean, as predicted by theoretical Ekman models [*Elipot and Gille, 2008*].

The third row of Fig. 3 shows the impact of near-inertial frequencies taken from  $\nu = -f/2\pi/2$  to 2 cpd. In Fig. 3g, the energy input rate in near-inertial frequency range (gray line, here multiplied by 10 for legibility) increases with latitude, peaks at 45° S and decreases further south. Similarly, in Fig. 3i, the variance of the near-inertial wind stress (gray, again multiplied by 10 for legibility) increases from 31° S to 45° S, then stays almost constant to 53° S, and finally decreases towards 59° S. The relative importance of the near-inertial to non-inertial frequencies (Fig. 3h) is quite modest, always less than 7%. The contribution from the near-inertial band is somewhat surprising. Why does the contribution from the near-inertial frequencies peak at latitudes to the north of the contribution from all the non-zero frequencies? There may be two reasons for this drop in the co-spectrum: first the low variance in ECMWF wind stress at higher latitudes [*Gille, 2005*], and second the drop of variance in the inertial band for drifter data towards higher latitudes because of the 12-hour interval used for computing drifter velocities which

approaches the local inertial frequency [*Lumpkin and Pazos*, 2007; *Elipot and Lumpkin*, 2008]. The existence of strong inertial oscillations in the Southern Ocean has recently been shown in the drifter data [*Chaigneau et al.*, 2008] that have not been regridded in time using the kriging procedure described in *Lumpkin and Pazos* [2007]. Thus the contribution from near-inertial frequencies is likely underestimated with the datasets used for this study.

## 4.2. Seasonal variability

The wind energy input rate estimates were recomputed after sorting the data between a summer season and a winter season. In general, the seasonal variability manifests itself latitudinally and also as a function of the frequency range considered.

From summer to winter, south of  $48^\circ$  S, the energy input rates from the mean (light blue and light red lines in Fig.4a) are not distinguishable within the error bars. North of this latitude the differences are statistically significant. The energy input rates from the non-zero frequencies (medium blue and medium red lines) increase by 69% on average from summer to winter. This increase also varies latitudinally, roughly decreasing from north to south, with almost twice the fractional seasonal difference at  $39^\circ$  S than at  $55^\circ$  S. As a consequence, since the total energy input (dark red and blue lines in Fig. 4a) is dominated by the non-zero frequencies, from summer to winter, the total input is increased on average by 67%, but this seasonal increase varies greatly: south of  $42^\circ$  S it averages to 42%, but north of  $42^\circ$  S it averages to 105%.

Although total energy input changes seasonally, the partitioning between anticyclonic and cyclonic components does not. Figs. 4d, e and f indicate no qualitative differences as a function of seasons.

The wind energy input in the near-inertial range (Fig. 4g) does show an increase from summer to winter, between 5% and 122% across the Southern Ocean. This pattern follows the trend of increased wind stress variance in winter in the near-inertial band, but it is less dramatic since the near-inertial variance of the wind stress can be multiplied by a factor up to 2.5.

### 4.3. Frequency decomposition of the input

For the purpose of discussing further the frequency decomposition of the energy input and comparing to previous estimates of the wind energy input, Fig. 2 shows the cumulative integration from  $|\nu| = 0$  to 2 cpd of the surface co-spectrum from data in the zonal band centered around  $41^\circ$  S for anticyclonic frequencies (dashed curves), cyclonic frequencies (thin solid curve) and both frequencies (heavy solid curve). The results in all other latitudinal bands are fairly similar to what is observed at  $41^\circ$  S.

Anticyclonic frequencies (dashed lines) contribute more to the total energy input than the cyclonic frequencies (thin solid lines), and this is true for all frequency ranges greater than zero. Again, this is due to a greater anticyclonic wind stress forcing reinforced by a greater anticyclonic response [Elipot and Gille, 2008]. This is a potential explanation for the predominance of anticyclonic “spin” in the Southern Ocean at these latitudes as revealed by the study of Griffa *et al.* [2008].

Time-varying components dominate the energy input relative to the zero-frequency contribution, as is also evident for all latitudes of the Southern Ocean in Fig. 3a. Motions at frequencies higher than  $(20)^{-1}$  cpd actually generally contribute to more than about 50% of the total. This is in contrast with the time scales at which potential energy is thought to increase via the work on geostrophic motions: Wunsch [1998] found in his estimate of

Eq. (1) that the contribution from the time mean accounted for about 94% of the global average value. However, the time-varying components were resolved up to approximately  $(20)^{-1}$  cpd only, the Nyquist frequency for T/P orbit repeat period, and contributions from higher frequencies, especially arising from wind fluctuations, were believed to be very small. *Hughes and Wilson* [2008] recomputed the wind energy input into geostrophic motions using updated altimetry data and scatterometer data, and their global integrals imply that the contribution of the time mean exceeds in fact 98% of the total.

*Wang and Huang's* (2004) estimates of the wind energy input to the Ekman layer, which varied approximately between  $12$  and  $20 \times 10^{-3} \text{ W m}^{-2}$  in the ACC region (see their Fig. 3), included only the contribution from frequencies up to  $|\nu| = 1/2$  cpd. For our estimates the contribution from this same frequency range (with an upper limit indicated by a vertical gray line in Fig. 2) are generally larger, here  $20.8 \pm 0.5 \times 10^{-3} \text{ W m}^{-2}$  at  $41^\circ \text{ S}$  in Fig. 2 but maximum with  $49 \pm 3 \times 10^{-3} \text{ W m}^{-2}$  at  $57^\circ \text{ S}$  (not shown). For their estimates, *Wang and Huang* used a theoretical model to compute the Ekman velocities at the surface. Their model assumes an infinitely deep ocean and a constant vertical viscosity. *Elipot and Gille* [2008] found this model to be the worst of 9 simple models that they tested. For the purpose of the wind energy input, the *Wang and Huang* model underestimates the co-spectrum because it requires that the velocities at the surface be directed  $45^\circ$  from the wind stress direction and therefore underestimate the energy input. Moreover, we find that higher frequencies from  $|\nu| = 0.525$  to  $2$  cpd, neglected by *Wang and Huang* [2004], add a non-negligible contribution, here about  $2.6 \pm 0.8 \times 10^{-3} \text{ W m}^{-2}$  at  $41^\circ \text{ S}$  or 11% of the total. Overall, the high-frequency contribution varies between 7% and 12% for the latitudes considered here. As mentioned previously, the variance of the

wind stress from numerical weather reanalyses as well as drifter data may be too low at high latitudes, and consequently the contribution from these frequencies may even be underestimated.

In the “near-inertial” frequency range, from  $\nu = -f/2\pi/2$  to 2 cpd, the anticyclonic frequencies input  $1.40 \pm 0.02 \times 10^{-3} \text{ W m}^{-2}$  at  $41^\circ \text{ S}$ , 6% of the total. Over the Southern Ocean as a whole, this percentage varies from 1.5% to 7% and is typically the same order of magnitude as the energy flux into wind-forced near-inertial mixed-layer motions over broad oceanic regions, as estimated from a slab-layer model [e.g. *Alford*, 2001]. These numbers imply that at least for the Southern Ocean, the wind energy is input to the Ekman layer predominantly at subinertial frequencies and not in the inertial band. However, the contribution of the near-inertial frequency range is likely underestimated in these datasets. *Elipot and Lumpkin* [2008] showed that much more energy is present in this band when the raw-drifter dataset is used compared to the historical 6-hourly kriged dataset. A definite answer to this uncertainty in the contribution of high frequencies at high latitude for the energy input rate in the Ekman layer is not likely to be accessible until higher temporal wind information (that is higher frequency sampling than 6 hours) is available.

## 5. Summary

A classic diffusion model, in which the turbulent stress is proportional to the vertical shear of the horizontal velocity, is used in the horizontal momentum balance equation of the ocean in the absence of large-scale pressure gradient in order to derive a spectral energy equation. In this framework, the real part of the cross-spectrum between the atmospheric stress upon the ocean and the surface ocean velocity is a measure as a function of frequency of the energy input “to the Ekman layer”. This energy is dissipated throughout the depth

of the oceanic boundary layer and may be passed below and is an important potential contributor to the mechanical energy budget of the ocean.

Surface drifter data, altimeter data, and reanalysis wind stresses are used to estimate first the zonally-averaged cross-spectrum between the wind stress and 15-m ageostrophic velocities for absolute frequencies of motions between 0 and 2 cpd, in the Southern Ocean. Because drifter velocities are not exactly representative of currents at the surface but at 15 m, some knowledge of the vertical structure of the transfer function is needed to obtain the cross-spectrum at the surface. The results from *Elipot and Gille* [2008] are used to infer the cross-spectrum at the surface and therefore to estimate the wind energy input to the Ekman layer in this region.

The spectral characteristics of this energy input are studied, notably its polarization. Results show that the combination of a stronger anticyclonic wind stress forcing associated with a greater anticyclonic response makes the contribution from the anticyclonic frequencies dominate the wind energy input, providing a potential explanation for the predominance of anticyclonic motions at these latitudes.

The latitudinal and seasonal variations of the wind energy input to the Ekman layer are closely related to the variations of the wind stress, for the mean and for the time-varying components. From these, the contribution from the near-inertial band follows a different trend, increasing from low latitudes to only about 45° S to decrease further south, possibly a consequence of the lack of variance in the drifter and wind stress data.

The uncertainties arising from the drifter data could potentially modify our quantitative results by up to 30%. However, we argue here that surface drifter velocities can be used to obtain global estimates of the wind energy input, provided that we obtain a good



understanding of vertical viscosity and boundary layer depth so that the surface cross-spectrum can be obtained from the 15 m cross-spectrum. This is of potential great importance as high frequency forcing and variability of the ocean is related to the sources of kinetic energy and near-inertial waves that are related to the mixing of the ocean and the global overturning circulation.

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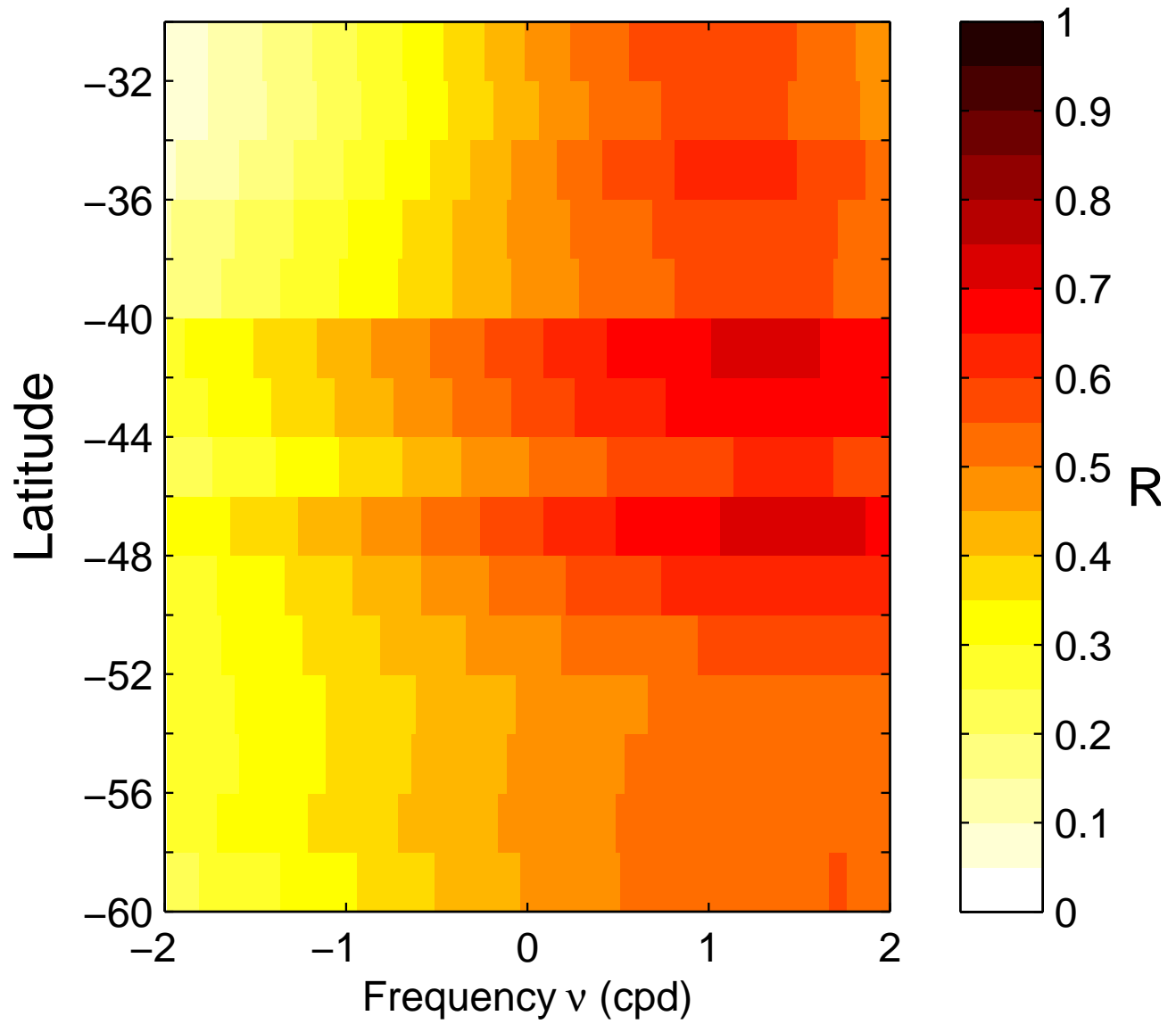
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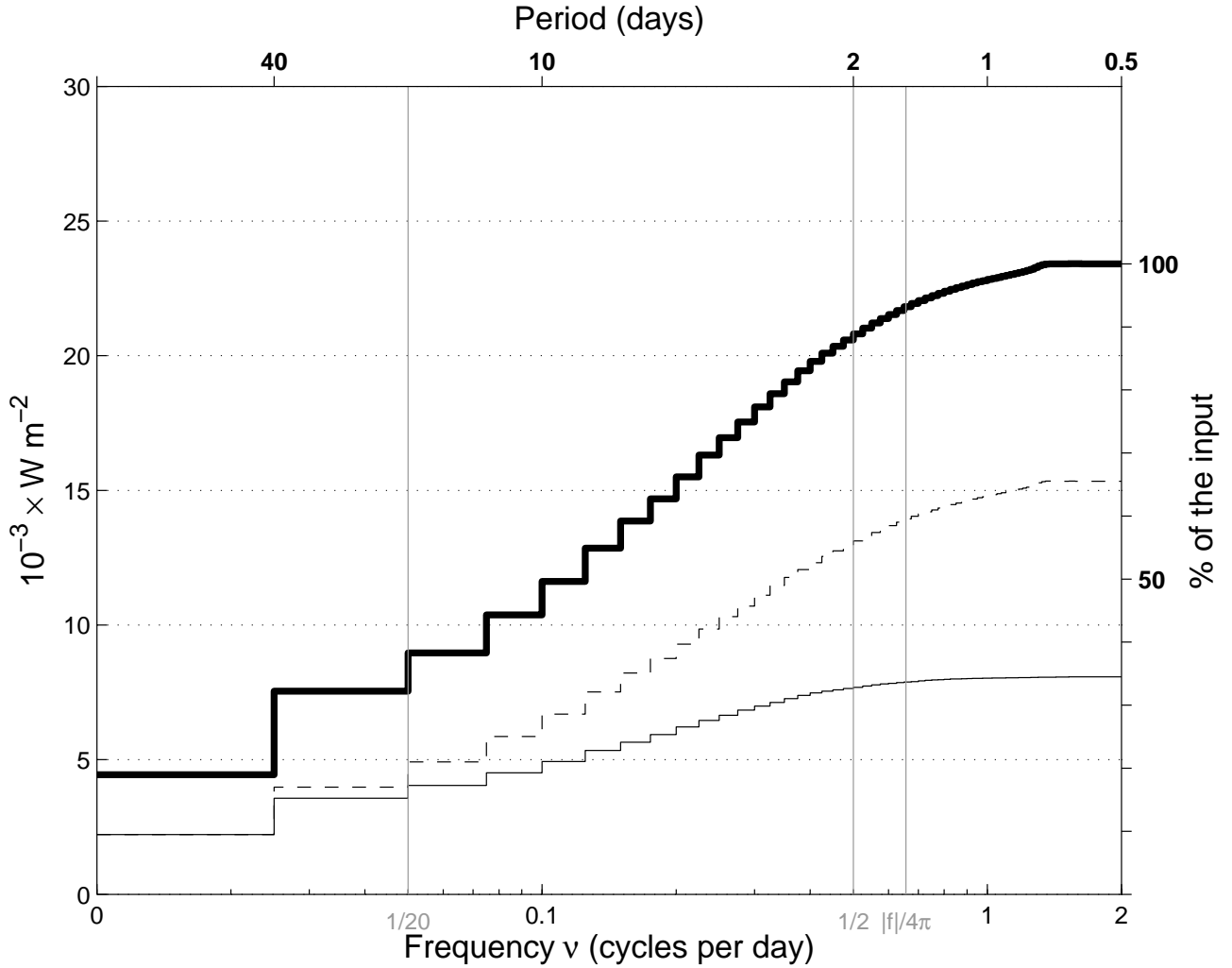
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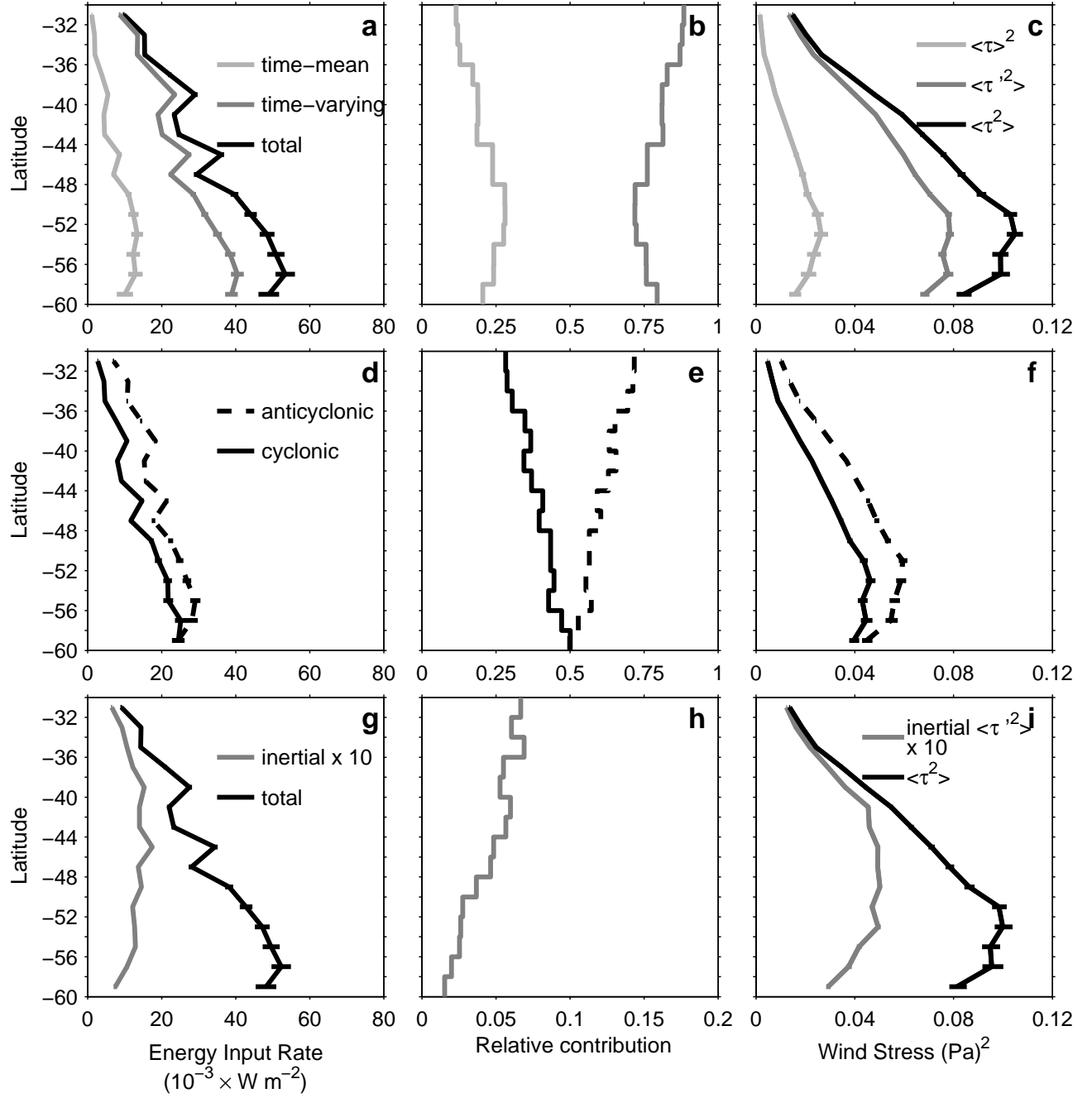


**Figure 1.** Ratio of the co-spectrum at 15 m to the co-spectrum at the surface.

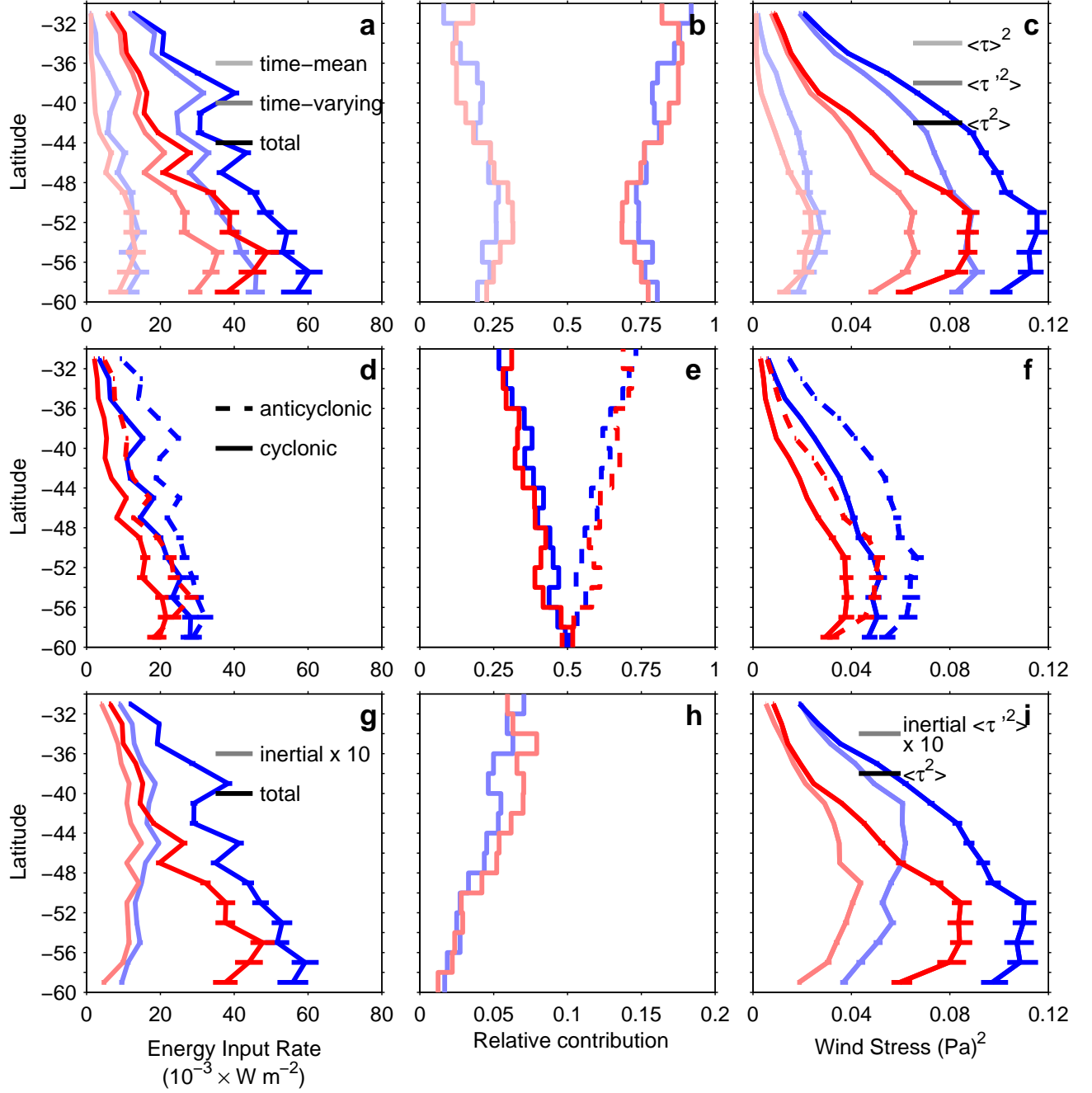


**Figure 2.** Cumulative integration of the co-spectra at the surface at  $41^\circ$  S for anticyclonic frequencies (dashed curve), cyclonic frequencies (thin solid curve) and the sum of anticyclonic and cyclonic frequencies (solid heavy curve). The axis of the abscissas is on a logarithmic scale but the lowest frequency plotted in each case is the zero frequency contribution, which is split evenly between the anticyclonic and cyclonic domains. Vertical lines mark specific frequencies.





**Figure 3.** Energy input rates across the Southern Ocean. (a) Contributions from the mean, the non zero frequencies and total within each latitudinal band. (b) Relative contributions of the zero frequency and of the non zero frequencies in the total energy input rate. (c) Mean value square, variance and mean square value of the wind stresses interpolated on the drifter positions. (d) Energy input rate contributions from the anticyclonic and the cyclonic frequencies. (e) Relative contribution of the anticyclonic and cyclonic frequencies for the total energy input rate. (f) Mean square value (variance plus half of the mean) of the wind stresses for anticyclonic and cyclonic frequencies. (g) Contribution from the near-inertial frequencies multiplied by 10 for legibility and contribution from the remaining frequencies. (h) Relative contribution of the near inertial frequencies for the total energy input rate. Note the change of scale for the abscissa compared to panels b and e. (i) Wind stress variance for the near-inertial frequencies multiplied by 10 for legibility and mean square value for the remaining frequencies. Error bars for the energy input rate are derived from the standard errors for the co-spectra. Error bars for the wind stress variance are derived from the formal 95% confidence intervals of the wind stress spectra.



**Figure 4.** Energy input rates across the Southern Ocean as a function of season. Blue shading: austral winter, red shading: austral summer. For the description of each panel see the caption of Fig. 3.